

Preconditioning a Parallel, Inexact Block-Jacobi Splitting of the S_N Algorithms

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Outline

- Introduction
- Problem formulation
- TSA
- Eigenvalue spectra
- Parallel performance and scaling

Introduction

- Parallel block-Jacobi S_N
 - Spatial domain decomposition
 - Finding efficient parallel sweep schedules for unstructured grids is difficult.
 - Parallel block-Jacobi algorithm [Yavuz and Larsen, 1989 and 1992]
 - Sub-domain interface angular fluxes are iterative unknowns.
 - Convergence degrades with increasing processors (N_p iterations needed in voids).
- Krylov methods
 - Can a Krylov iterative method improve efficiency?
 - Can Transport Synthetic Acceleration (TSA) be used as an effective preconditioner?

Problem Formulation

- Time-independent S_N equations
 - Steady-state, isotropic scattering
 - Quadrature $\{\mu_m, w_m\}, m = 1, \dots, N_A$
 - Linear discontinuous finite element spatial discretization on cells
 $j = 1, \dots, N_c$

$$(\hat{\Omega}_m \cdot \nabla + \sigma_t(r))\psi_m(r) = \frac{1}{4\pi} \sigma_s(r)\phi(r) + q_m(r), \quad \phi(r) = \sum_{m=1}^{N_A} w_m \psi_m(r)$$

$L\Psi$

$MS\Phi$

Q

$D\Psi$

Problem Formulation

- Operator Notation
 - Define the vector $\Psi \in \mathbb{R}^{2N_c N_A}$, $\Phi \in \mathbb{R}^{2N_c}$

$$L\Psi = MS\Phi + Q$$

$$\Phi = D\Psi$$

$$\Phi = DL^{-1}MS\Phi + DL^{-1}Q$$

$$(I - DL^{-1}MS)\Phi = DL^{-1}Q$$

Problem Formulation

- Parallel block-Jacobi iteration
 - Spatially decomposed problems
 - Inverting L on every sub-domain p independently
 - Incoming boundary angular fluxes are specified from previous iteration or initial guess.

$$\Psi_p = L_p^{-1} \left(M_p S_p \Phi_p + \sum_{p'} L_{p,p'} \Psi_{p'} + Q_p \right)$$

$$\Phi_p = D_p \Psi_p$$

- L_p interior angular fluxes
- $L_{p,p'}$ couples sub-domain p to adjacent sub-domain p' .
- Scalar fluxes and boundary angular fluxes are iterative unknowns.

Problem Formulation – Parallel Implementation

- Only outgoing angular fluxes on sub-domain boundaries are transferred.
- Define $\Psi_{p,p'} = W_{p,p'}^T \Psi_p$ for sub-domain p' adjacent to p
- $W_{p,p'}^T$ extracts the outgoing angular fluxes on the boundary from Ψ_p .
- $W_{p,p'}$ maps back onto the full angular flux vector.
- Write the iteration in terms of an “augmented vector”

$$\begin{bmatrix} \Phi_p \\ \Psi_{p,p'} \end{bmatrix}$$

Problem Formulation

- Source iteration

$$\begin{bmatrix} \Phi_p \\ \Psi_{p,p'} \end{bmatrix}^{l+1} = \left(\begin{bmatrix} D_p \\ W_{p,p'}^T \end{bmatrix} L_p^{-1} \left[M_p S_p \left(-\sum_{p'} L_{p,p'} W_{p,p'} \right) \right] \right) \begin{bmatrix} \Phi_p \\ \Psi_{p,p'} \end{bmatrix}^l + \begin{bmatrix} D_p \\ 0_{p,p'} \end{bmatrix} L_p^{-1} Q_p$$

- Krylov iterative method

$$\left(I - \begin{bmatrix} D_p \\ W_{p,p'}^T \end{bmatrix} L_p^{-1} \left[M_p S_p \left(-\sum_{p'} L_{p,p'} W_{p,p'} \right) \right] \right) \begin{bmatrix} \Phi_p \\ \Psi_{p,p'} \end{bmatrix} = \begin{bmatrix} D_p \\ 0_{p,p'} \end{bmatrix} L_p^{-1} Q_p$$



Transport Synthetic Acceleration

- Source iteration is Richardson iteration.

$$\Phi^{l+1} = DL^{-1}MS\Phi^l + DL^{-1}Q$$

- Preconditioning the Richardson iteration to improve convergence. This equivalent to synthetic acceleration.

$$\Phi^{l+1} = (I + PS)DL^{-1}MS\Phi^l + (I + PS)DL^{-1}Q$$

$$P = URV, R \approx (L - MSD)^{-1}$$

- U and V are appropriate projection and interpolation operators.

Transport Synthetic Acceleration

- For Diffusion Synthetic Acceleration (DSA), R is calculated by solving a diffusion equation.
- For Transport Synthetic Acceleration (TSA), R is simply a low-order transport equation.

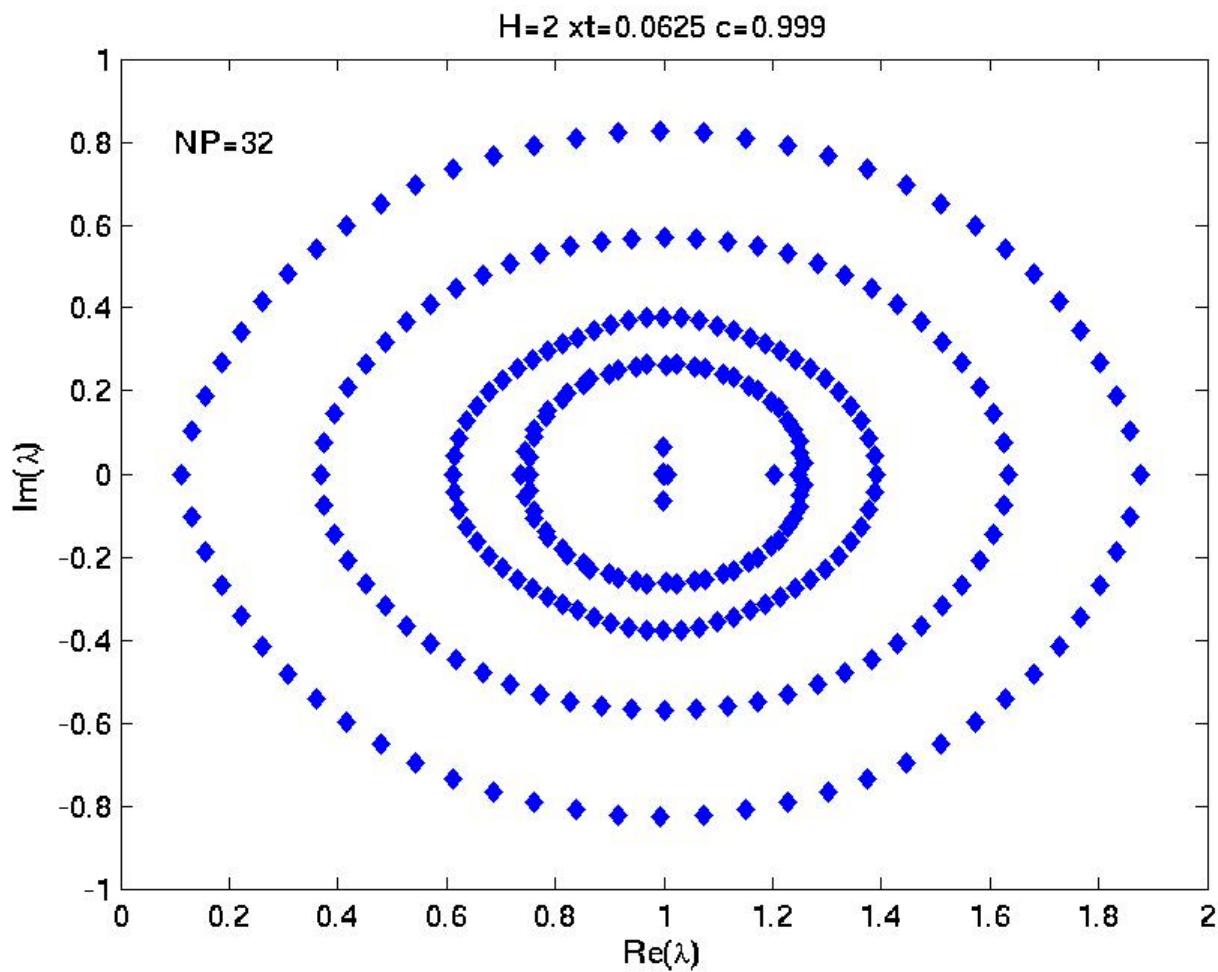
$$\left(\hat{\Omega}_m \cdot \nabla + \sigma'_t(r) \right) \psi'_m(r) = \frac{1}{4\pi} \sigma'_s(r) \phi'(r) + q'_m(r)$$

- In traditional TSA: $\sigma'_s = (1 - \beta) \sigma_s$ and $\sigma'_t = (\sigma_t - \beta \sigma_s)$
 - $\beta = [0, 1]$
 - L and S are modified in the low-order equations to satisfy same conservation properties as the high-order problem.
 - Typically, low-order equations use lower order quadratures and less scattering.

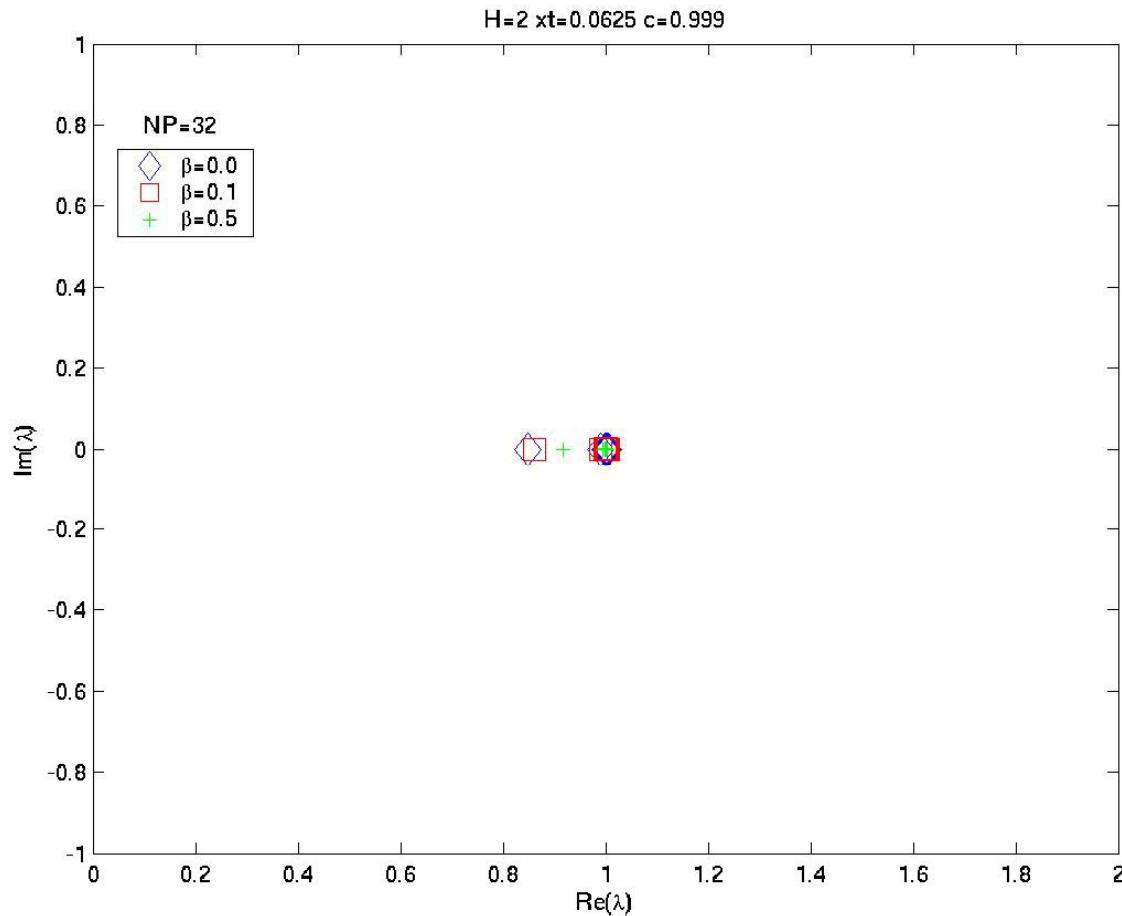
Transport Synthetic Acceleration

- Modified TSA: $\sigma'_s = \beta\sigma_s$ $\sigma'_t = \sigma_t$
 - σ_t remains the same and S is reduced in the low-order equations.
 - Always convergent
 - Simpler implementation within the time and energy dependent calculations
 - Can also use lower order quadratures

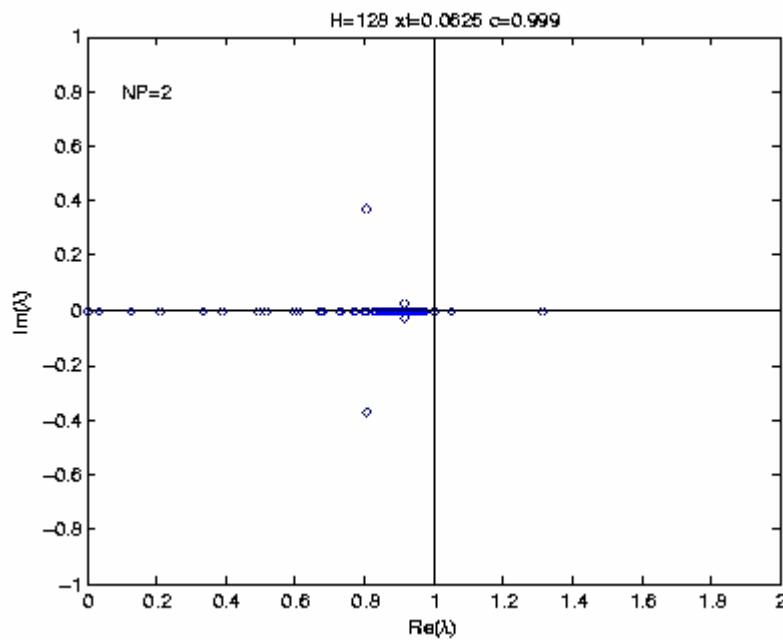
Eigenvalue Spectra for Source Iteration



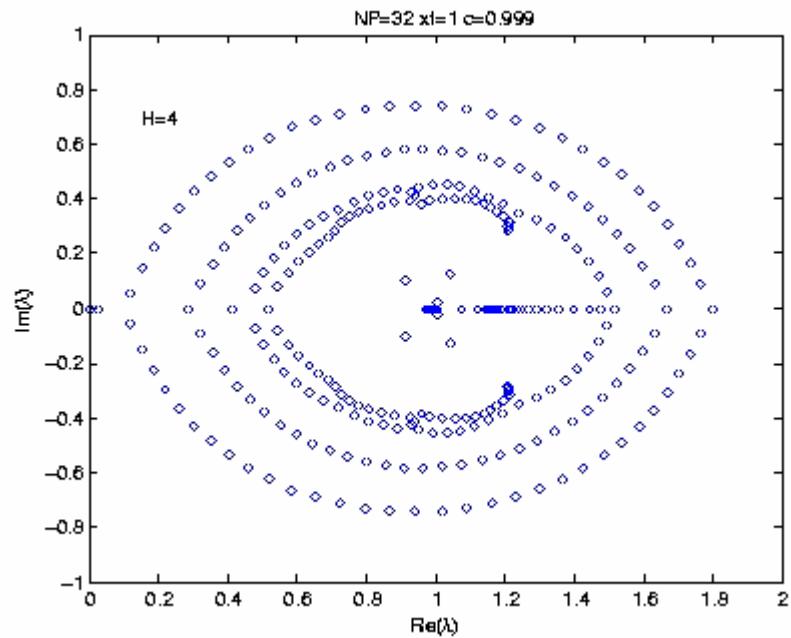
Eigenvalue Spectra for Modified TSA



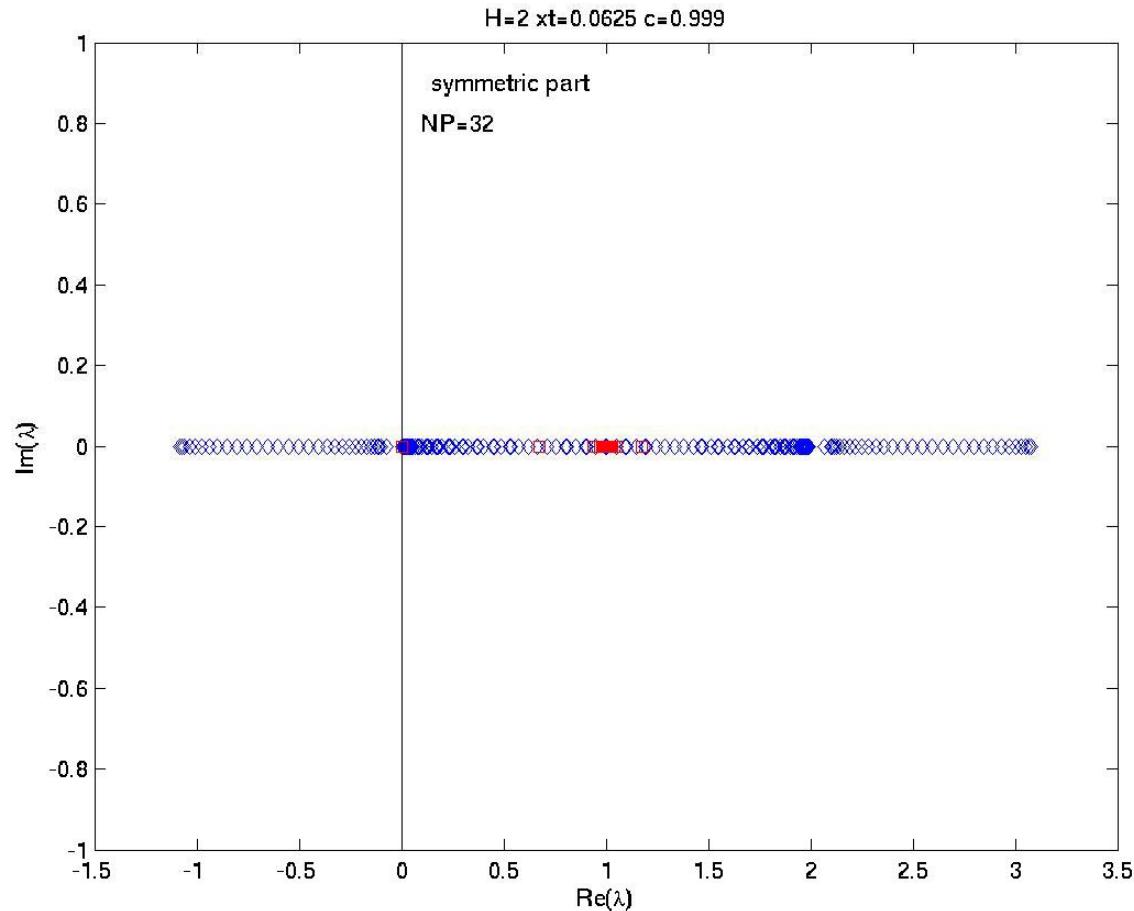
Eigenvalue Spectra for a Fixed Physical Problem (H constant) as a Function of Number of Processors



Eigenvalue Spectra for a Fixed Number of Processors as a Function of Thickness



Symmetric Part of A



Linear Algebra Observations

- A is non-normal
- For $(I - DL_0^{-1}L_b)\Phi = DL_0^{-1}Q$
 - d = degree of minimal polynomial, makespan of the subdomain graph
= N_p for 1D
 - = $2\sqrt{N_p}$ for 2D
 - = $3\sqrt[3]{N_p}$ for 3D
- For GMRES(m), when $m \geq d$, it converges, otherwise, stagnation is possible since A is indefinite.

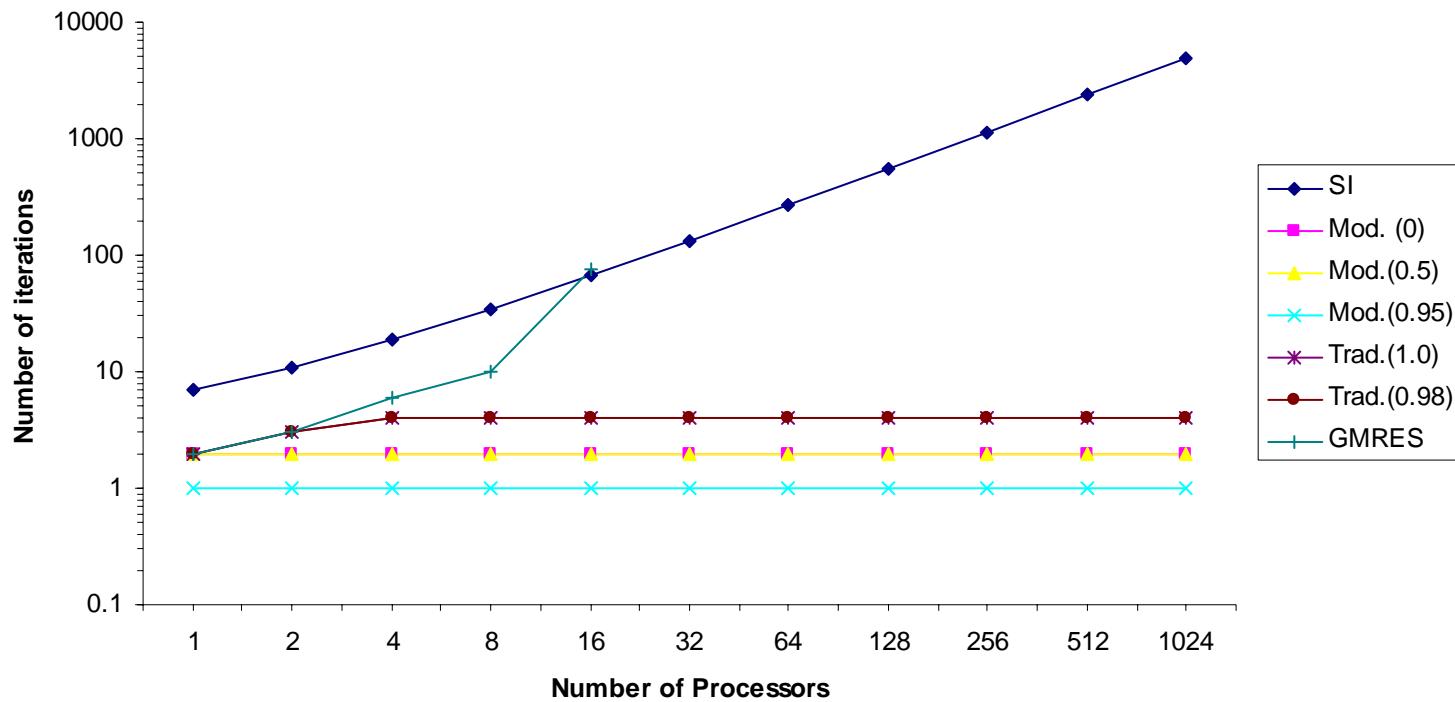
Parallel Performance and Scaling

- Problem Characteristics
 - S_8 quadrature
 - Fixed problem width $H = 2$ cm (equal-width mesh spacing)
 - Constant isotropic distributed source $q = 1$
 - Constant material properties $\sigma_t = 1/16 \text{ cm}^{-1}$, $c = 0.999$
 - Vacuum boundary conditions
 - Relative convergence criteria: 10^{-5} (high-order) and 10^{-7} (low-order)
 - Compare GMRES(15), source iteration, traditional TSA ($\beta=1.0$ and $\beta=0.9$), and modified TSA ($\beta=0.0$, $\beta=0.5$ and $\beta=0.95$)

Parallel Performance and Scaling

Weak Scaling

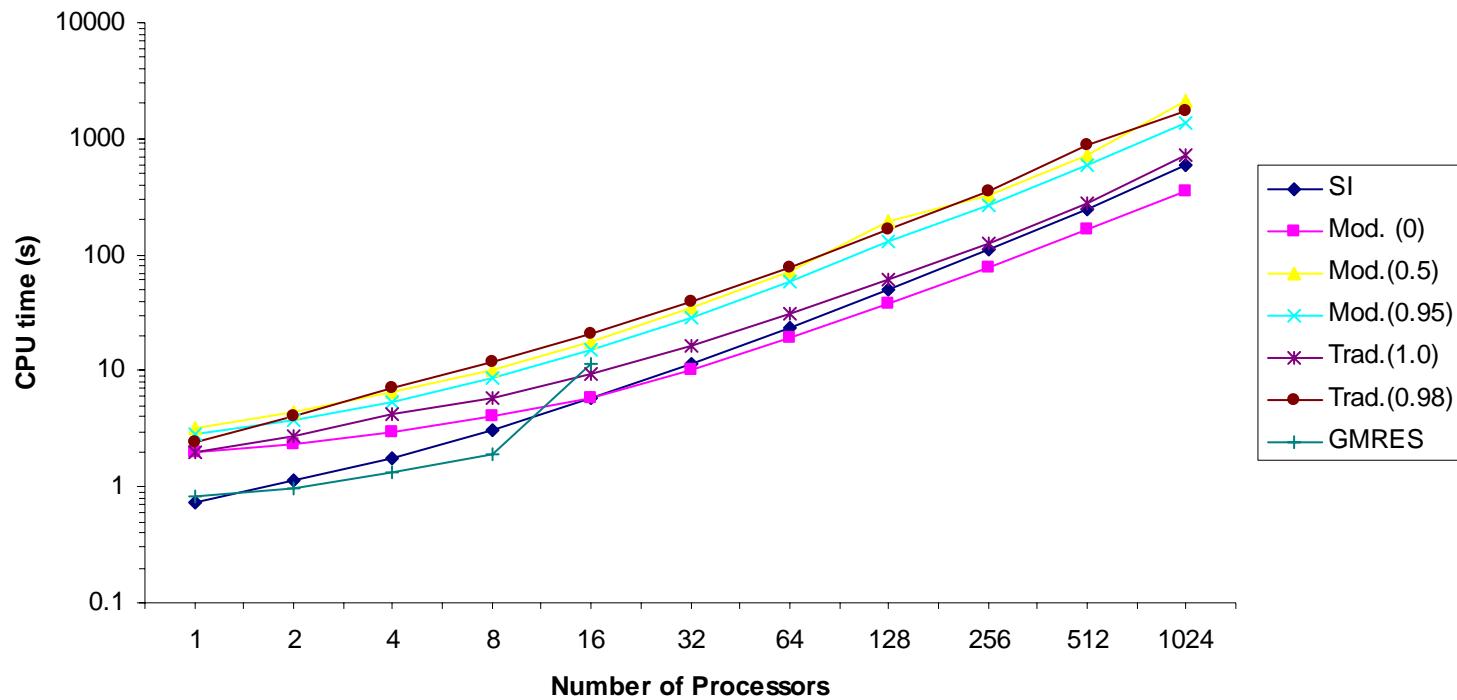
- Fixed work per processor: $n = 65536 \times N_p$



Parallel Performance and Scaling

Weak Scaling

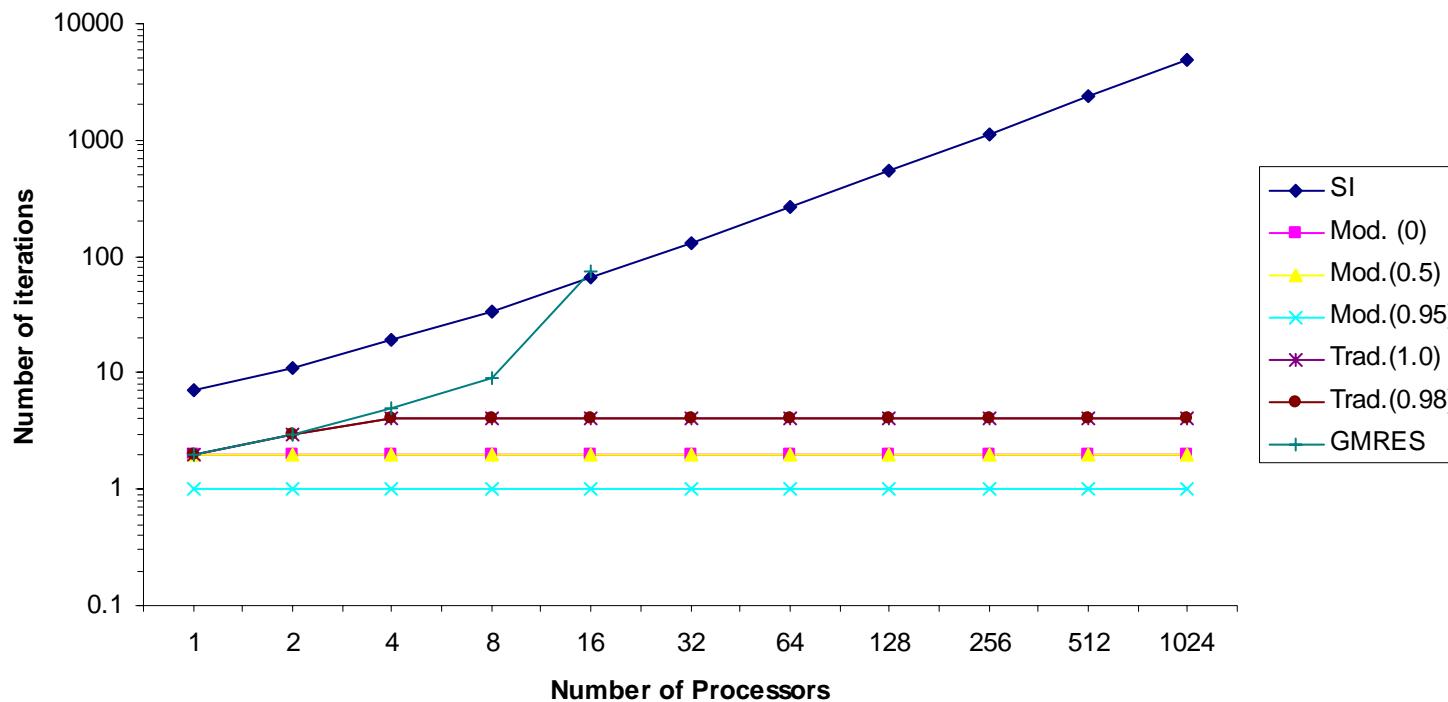
- Fixed work per processor: $n = 65536 \times N_p$



Parallel Performance and Scaling

Strong Scaling

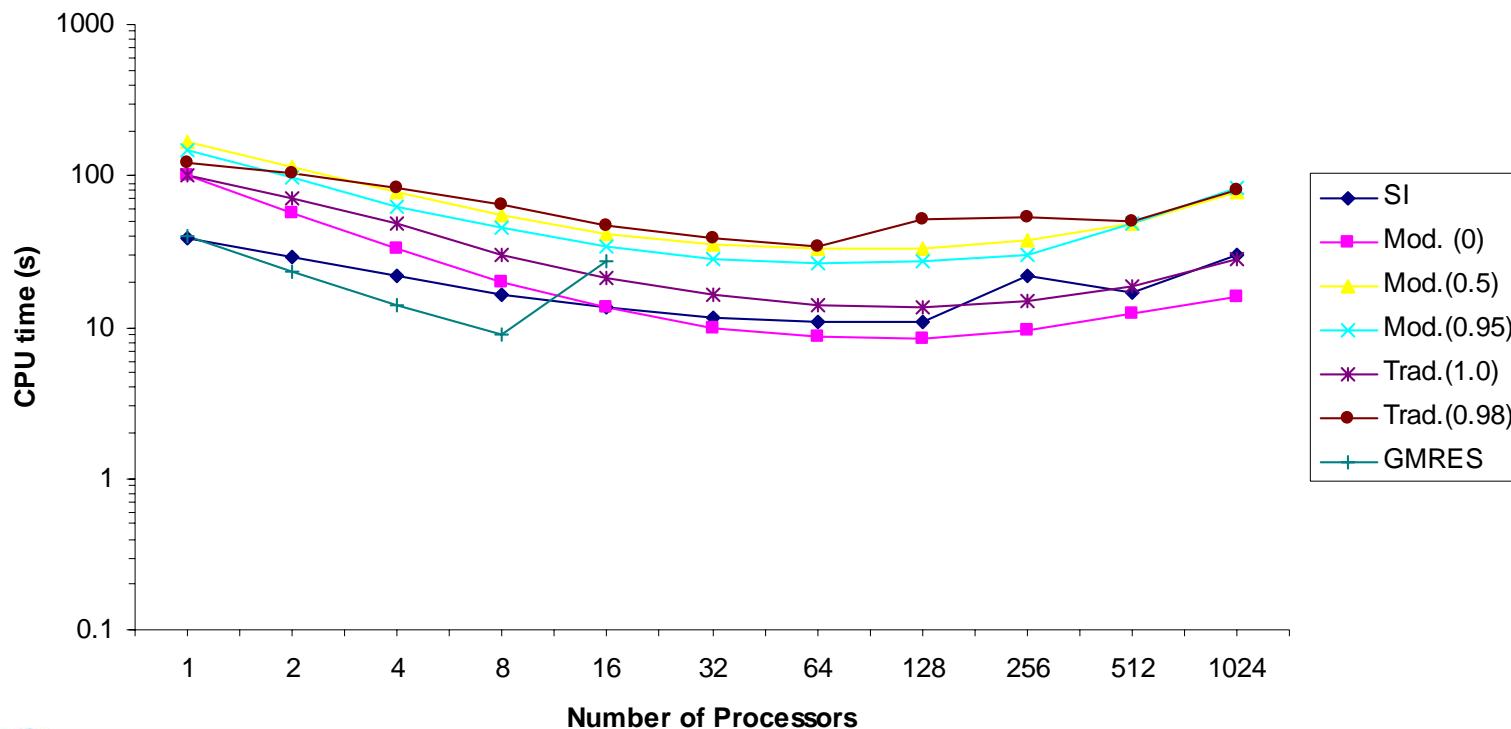
- Fixed problem size: $n = 2097152$



Parallel Performance and Scaling

Strong Scaling

- Fixed problem size: $n = 2097152$



Summary and Conclusions

- GMRES stagnates for optically thin problems when the restart value is less than the number of processors.
- Modified TSA is an effective preconditioner for GMRES when the problem is optically thin.
- Need to explore 2D problems